

# ENABLING DIGITAL TECHNOLOGIES FOR IRON AND STEELMAKING PROCESSES

***Drs. Chenn Zhou and Tyamo Okosun***

Center for Innovation through Visualization and Simulation (CIVS)  
Steel Manufacturing Simulation and Visualization Consortium (SMSVC)

Purdue University Northwest, Hammond, IN

[civs@pnw.edu](mailto:civs@pnw.edu)

*“Where Ideas Become Reality”*



CENTER FOR INNOVATION THROUGH  
VISUALIZATION & SIMULATION

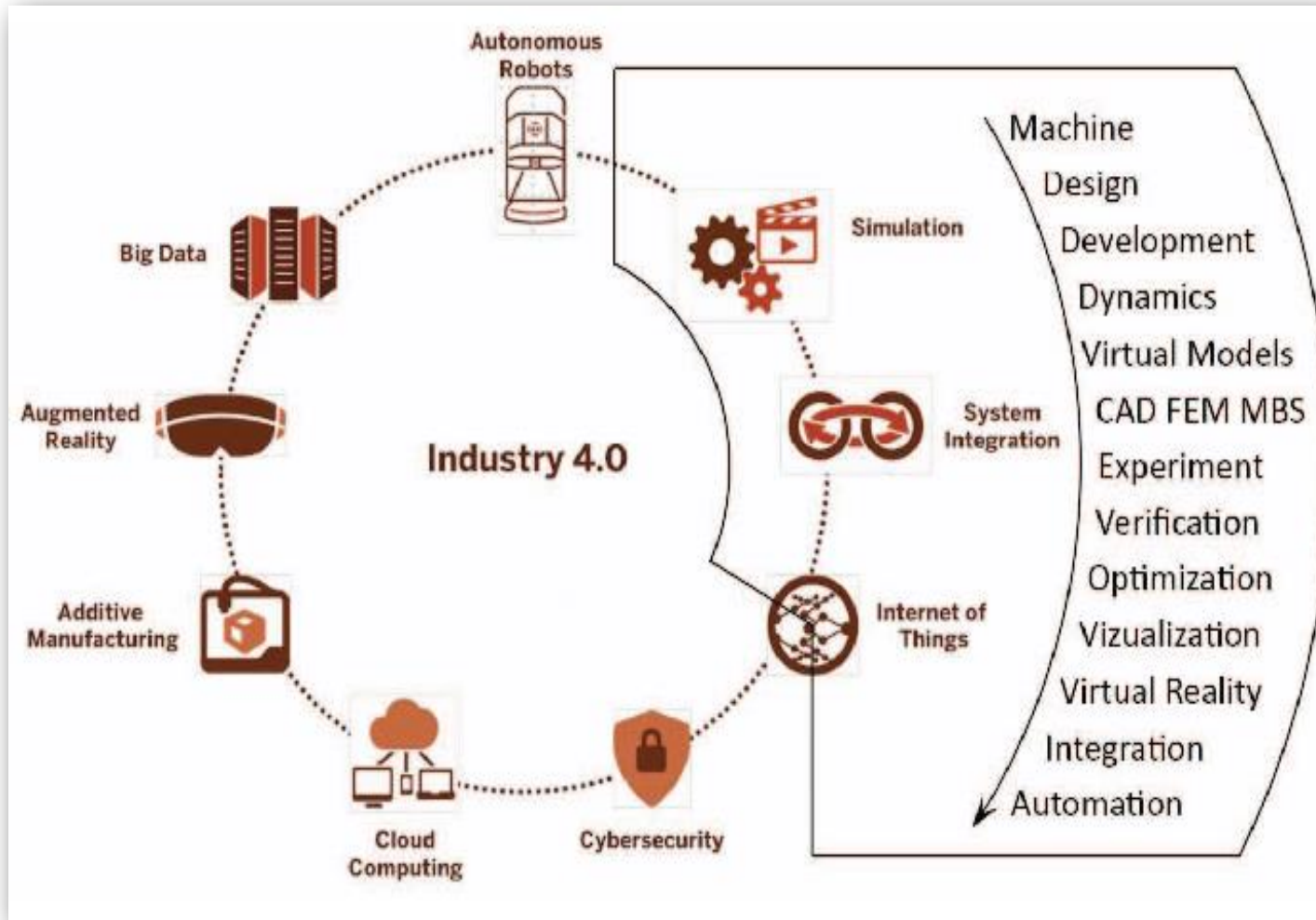
**PURDUE**  
UNIVERSITY  
**NORTHWEST**

2200 169th Street  
Hammond, IN 46323

219.989.2765  
[civs@pnw.edu](mailto:civs@pnw.edu)  
[pnw.edu/civs](http://pnw.edu/civs)



[steelconsortium.org](http://steelconsortium.org)



- **Simulation** such as **Computational Fluid Dynamics (CFD)**: Physics-based predictions
- **Visualization (VR & AR ...)**: Effective communication and monitoring
- **High Performance Computing (HPC)**: Accelerate computations
- **Machine Learning (ML)**: Predictions based on historic data (a branch of Artificial Intelligence)
- **Digital Twin**: Virtual replicas of physical devices or processes

Jiri Kovar et. al., "Virtual Reality in Context of Industry 4.0", [2016 17th International Conference on Mechatronics - Mechatronika \(ME\)](#)

# CIVS (since 2009) & SMSVC (since 2016)

## Center for Innovation through Visualization and Simulation (CIVS)

### ➤ Missions

- Innovation
- Application (**\$40+ million** savings from 5 out of 460+ projects)
- Education

### ➤ Strategies

- Integration of technologies
- Application driven
- Partnerships (**160+** organizations)



## Steel Manufacturing Simulation and Visualization Consortium (SMSVC)

### Research Areas

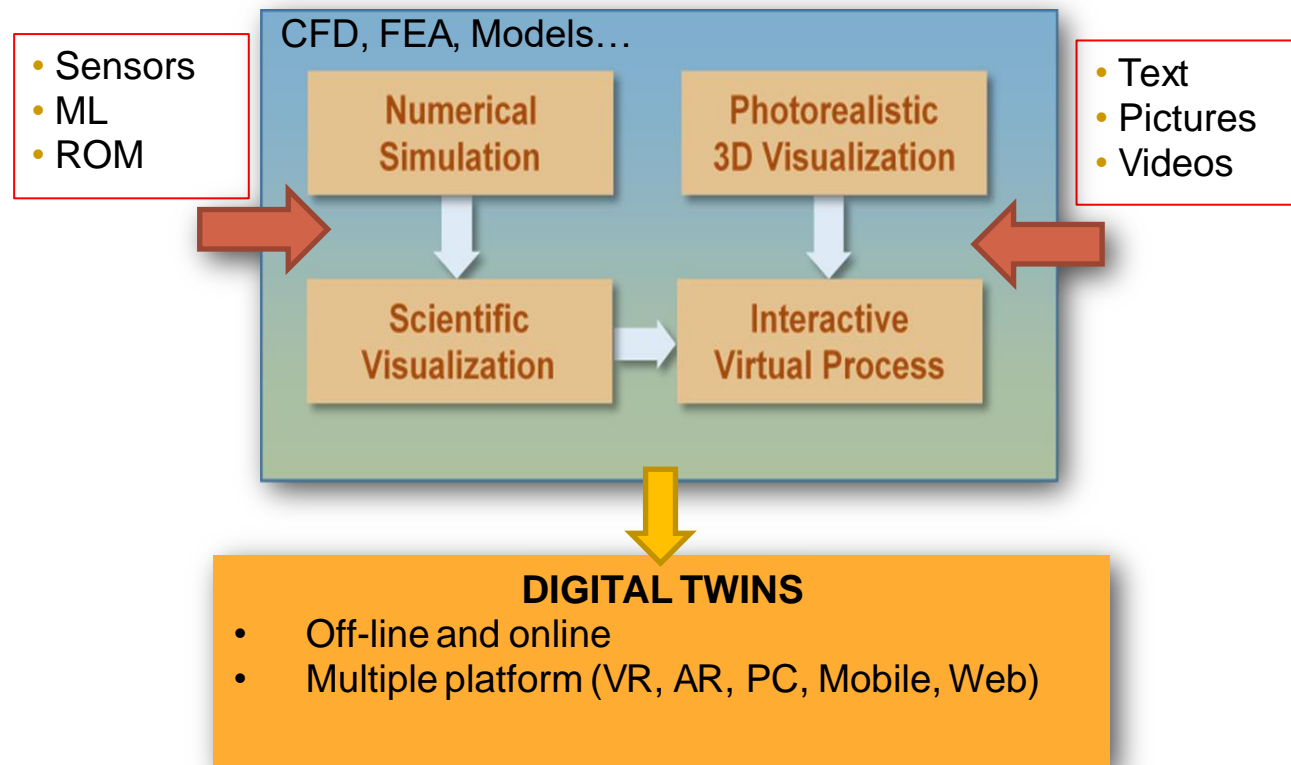
- 1) Energy Efficiency
- 2) Environmental Impacts
- 3) Operation Efficiency
- 4) Raw Materials Utilization
- 5) Reliability and Maintenance
- 6) Smart Manufacturing
- 7) Workforce Development
- 8) Workplace Safety

### Project Examples

- Blast Furnace
- Electric Arc Furnace
- Caster Digital Twins
- Smart Ladle
- Spray Cooling Simulator
- Reheating Furnace
- Hydrogen combustion

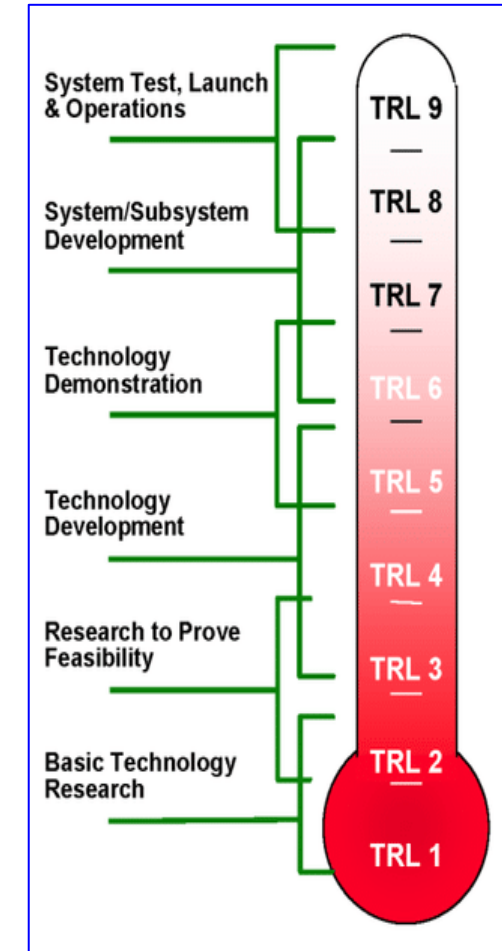


# METHODOLOGY AND BENEFITS



## Benefits

- **Insight:** visualizing flow fields
- **Foresight:** predict “future” or “what if”
- **Efficiency:** speed-up emerging & transformative technology R&D in all stages (TRL1 – 9\*)



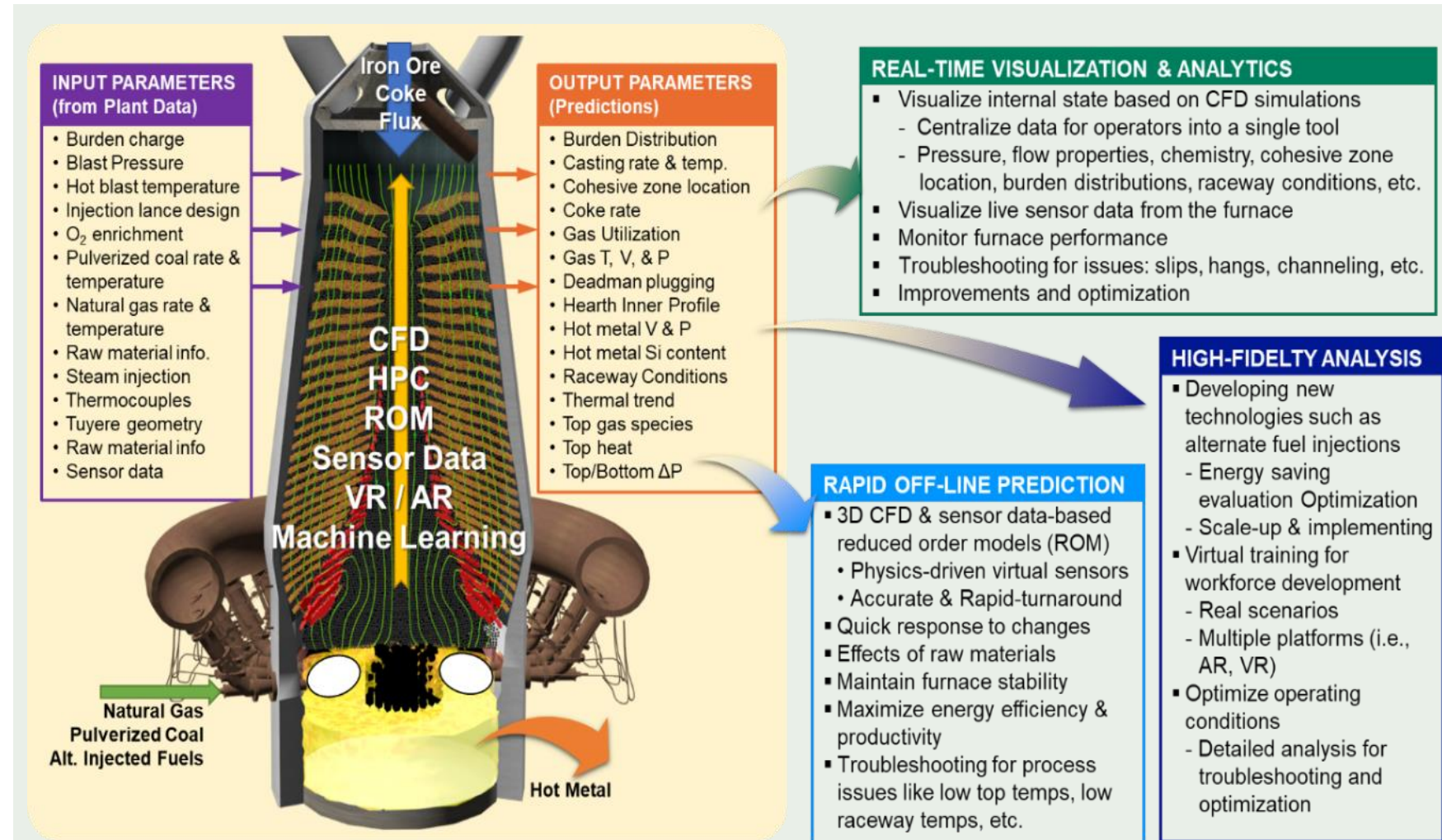
\*Felici, Massimo. (2015). SECurity and trust COoRDination and enhanced collaboration Deliverable 6.3 Identification of Future Emerging Issues/Topics (Final Version).



## ➤ Efficient, effective, and economical

- Physics-based, data driven decisions
- Lower risk & shorter cycle for developing and implementing transformative technologies
- Real time monitoring and control
- More effective training
- Better communication

***The IVBF Platform will be applicable for many other iron & steelmaking processes***



A proposed **Next Generation Integrated Virtual Blast Furnace**  
(led by Dr. Tyamo Okosun at PNW with industry, national lab, and universities)

# VIRTUAL BLAST FURNACE

## ➤ Issues:

- Energy efficiency
- Pollutant Emissions
- Campaign life
- Downtime
- Training

## ➤ Major Outcomes (since 2002):

- Virtual blast furnaces used for
  - Design
  - Troubleshooting
  - Optimization
  - Scale-up
  - Training
- Improvement of energy efficiency and emissions
- Downtime reduction

## ➤ Sponsors and Partners:

- AIST, AISI, ArcelorMittal, Cleveland-Cliffs, DOE, Stelco, SMSVC, Tata Steel, Union Gas, and U.S. Steel



CFD + FEA + HPC + Visualization

### Carbon Footprint Impact

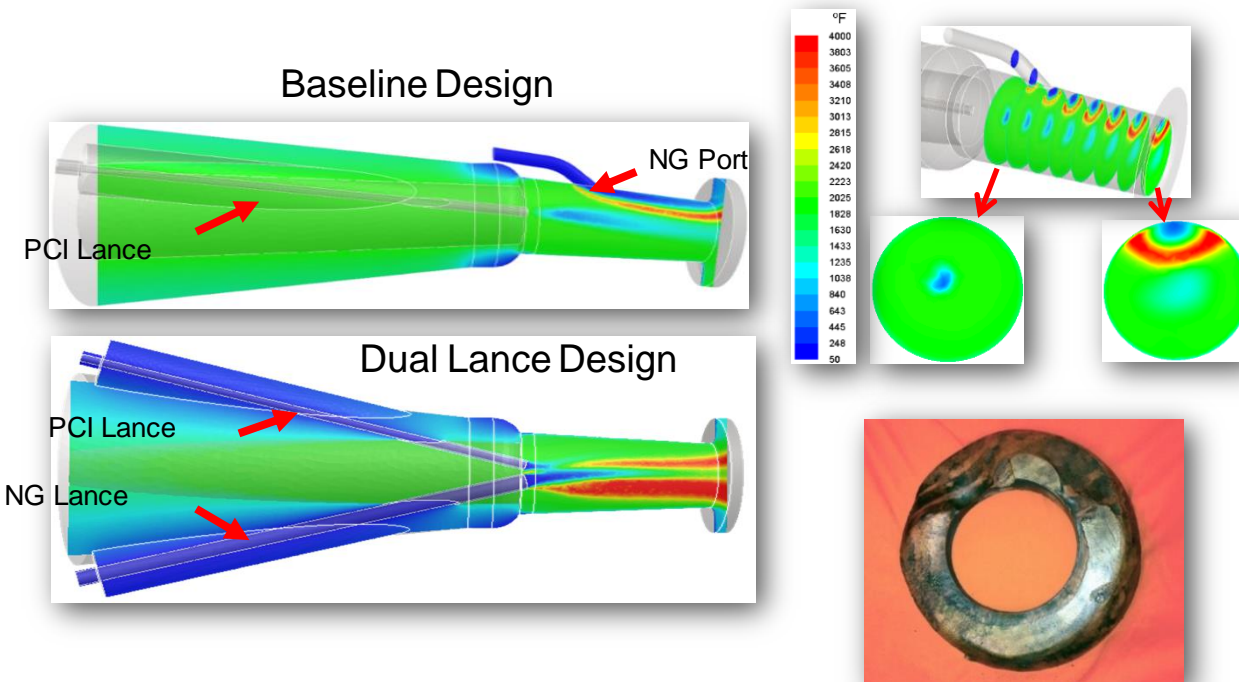
- For 4.5% energy savings, around 120,000 tons of CO<sub>2</sub>/year reduction per a typical blast furnace (BF)
- For 10% energy savings, around 260,000 tons of CO<sub>2</sub>/year reduction per a typical BF and 4.5 million tons of CO<sub>2</sub>/year reduction across the entire U.S. steel industry



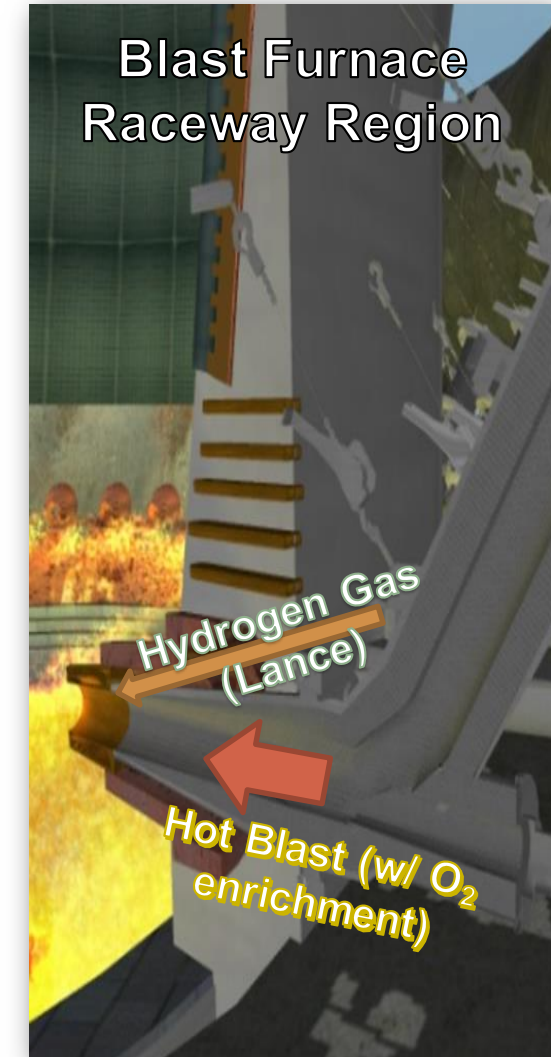
# OUTCOME EXAMPLES

## Cleveland-Cliffs Blast Furnaces

- **Design:** Implemented dual lance design - increased coal combustion efficiency by **25%**
- **Optimization:** Potential improvement of **2.5%** productivity using NG as PCI carrier gas
- **Troubleshooting:** Identified causes of tuyere failures



- **H<sub>2</sub> Injection**
  - Reduce CO<sub>2</sub>
  - instability due to quenching effect
- **Premixed NG + H<sub>2</sub>**
  - Boost H<sub>2</sub> rate
  - Balance injections to achieve desired flame T and coke rate with lower CO<sub>2</sub>
- **Preheated H<sub>2</sub> Injection**
- **Preliminary results:**
  - Fundamental understandings
  - Highest stable injection rates
  - Quantitative correlations



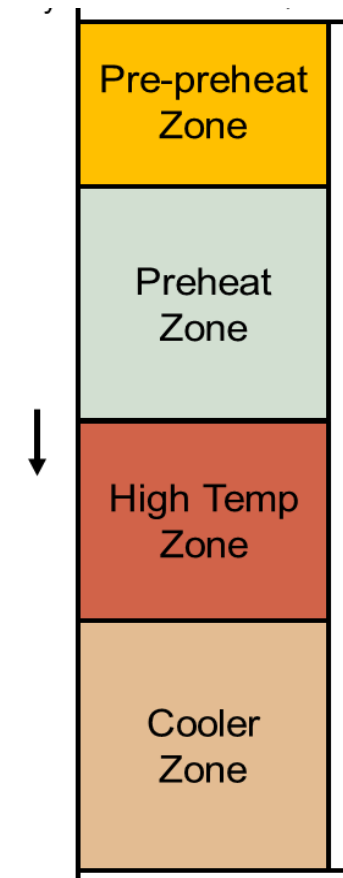
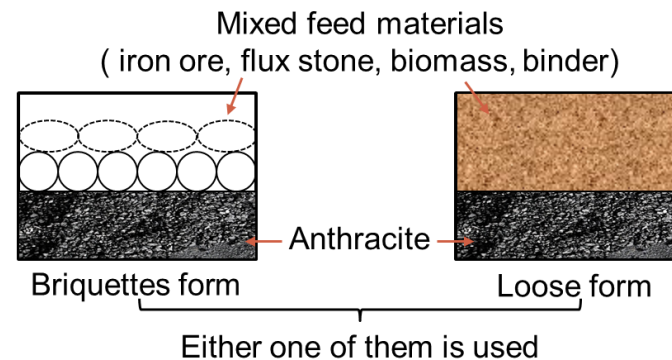
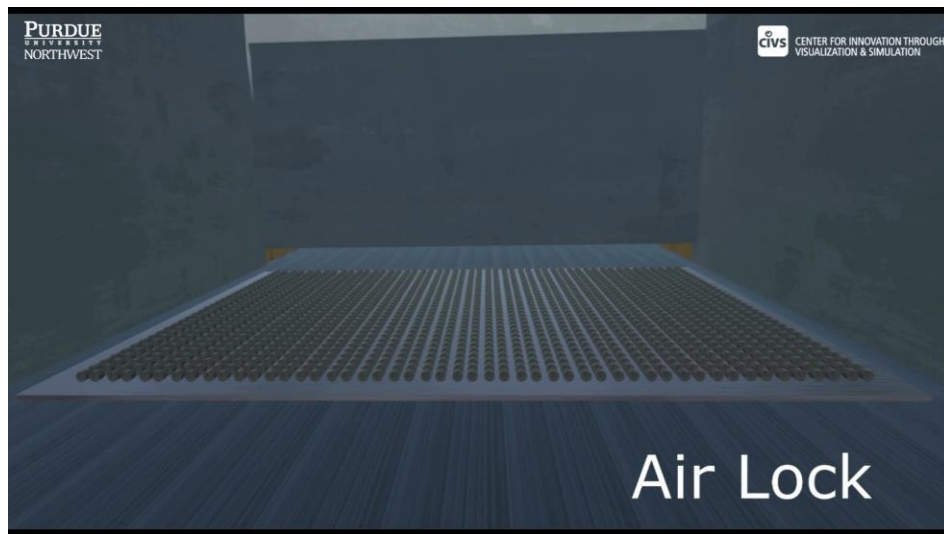
# E-IRON NUGGET PROCESS SCALE-UP

## ➤ Process:

- A novel iron smelting process developed by Carbontec Energy Corporation (CarbonTec) to utilize biomass to replace coke
  - To produce high quality pig iron nuggets
  - To significantly reduce CO<sub>2</sub> emissions

## ➤ Objective (12 months):

- Utilizing High Performance Computing (HPC) Computational Fluid Dynamics (CFD) simulations to guide furnace designs for the E-Iron™ Nugget process



Mixing iron ore or steel mill wastes with renewable biomass and a lime flux to produce a self-reducing pellet



# PROJECT OUTCOMES

## ➤ Research and development (TRL 4&5):

- Developed CFD Models for predicting the success or failure of iron reduction and nugget formation (13 reactions)
- Validated with **pilot scale data** with 4 cases (diff. < 5%)

## ➤ Scale-up (commercialization demonstration) (TRL 7&8)

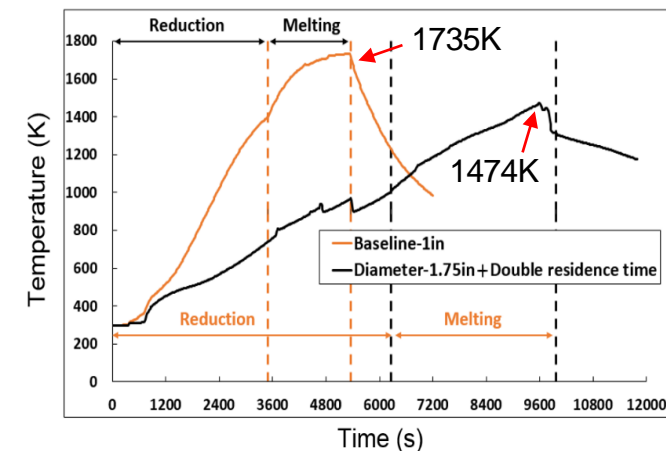
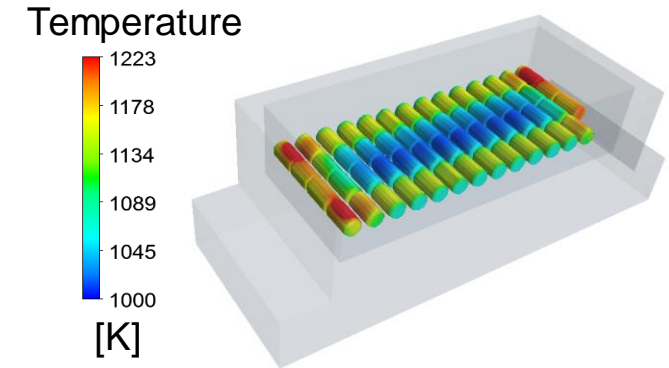
- Recommendations for the final design of the production scale furnace (100,000 tonnes/yr.)
- 15 iterations with 6 variables: temperatures in preheating, reduction, and melting zones; residence time in reduction & melting zones; and briquette size

## ➤ Visualization

- Visualization of a plant in operation
- CFD data integrated to show process details and highlight ways to increase production and lower production costs.

## ➤ Benefits:

- CFD simulations verified the technology, provided better understanding, and assisted in the final design to avoid pitfalls
- Visualization helped for marketing and convincing investors
- Three lines (100,000 tonnes/year/line) to be built in Burns Harbor, Indiana



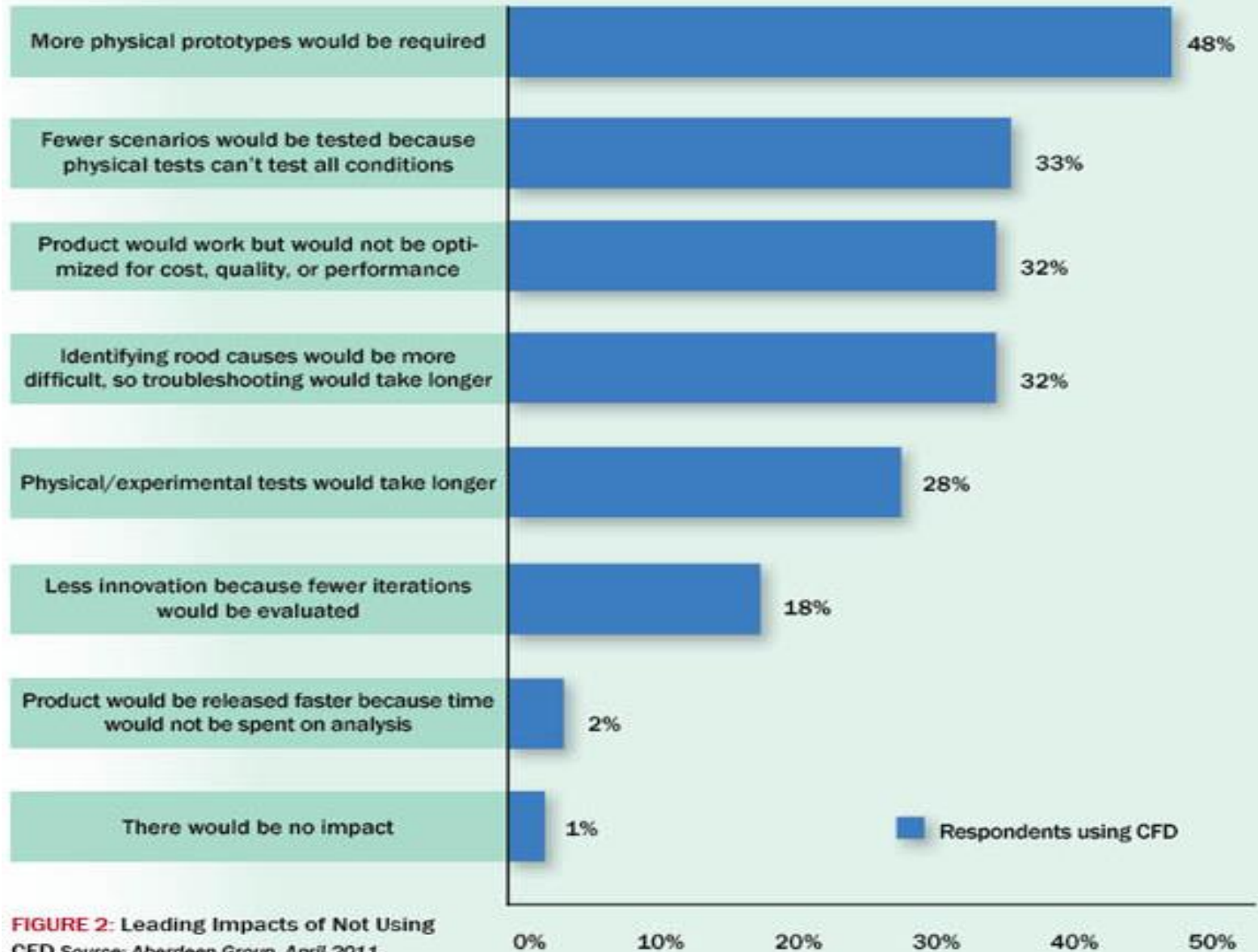
- Benefits: Speed up computational time with more details
- HPC availability: national labs, universities, companies, & cloud
- **DOE AMO HPC for Manufacturing (HPC4Mfg) Program**
  - Two regular solicitations annually, fall & spring
  - Eligibility: U.S. manufacturers
  - Submission: 1) concept paper; 2) full proposal
  - Budget: \$300,000 for phase 1 to support computing cycles and work performed by National Labs, universities, and non-profit partners.
  - Cost share: 20% from industry
  - National Labs: LLNL, ANL, ORNL
  - PNW CIVS & SMSVC Role:
    - Subcontractor of national labs
    - Participated in 6 projects

## Example

### Utilizing HPC to Model the E-Iron Nugget

- CarbonTec
  - Principal Investigator to submit proposals and sign CRADA with LLNL
  - Provide geometry and operating conditions
  - Implement results
- LLNL
  - HPC facility
  - Selection of experimental data for validation
  - Verification of chemical kinetics
  - Review models and results
- PNW CIVS
  - Subcontractor from LLNL
  - Developed and validated CFD Models
  - Run CFD simulations on LLNL cluster
  - Create visualization simulators

# DIGITAL TECHNOLOGIES TO ACCELERATE DEVELOPMENT PROCESS

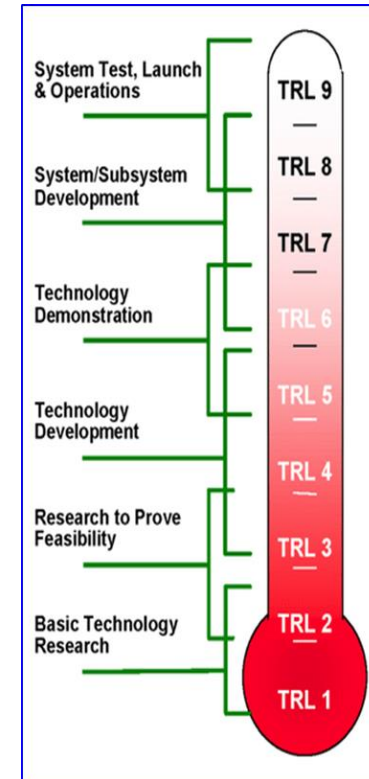


<https://www.digitalengineering247.com/article/the-role-of-cfd-in-product-development/>



# DIGITAL TECHNOLOGIES TO ACCELERATE DEVELOPMENT PROCESS

- **TRL 2:** Technology concept formulated (**CFD**) - Refine new concept, evaluate feasibility
- **TRL 3:** Experimental proof of concept (**CFD**) - Complementary to lab scale prototype, identify key parameters, verification of the proof of concept
- **TRL 4:** Technology validated in lab (**CFD**) - Complementary to small-scale prototype integrated with complementing subsystems, validation of the new technology
- **TRL 5:** Technology validated in relevant environment (**CFD, Visualization**) - Complementary to large scale prototype integrated with supporting components, assess parameters for scale-up
- **TRL 6:** Technology pilot demonstrated in relevant environment (**CFD, Visualization**) - Assist in demonstrating the technology as fine-tuned to a variety of operating conditions in relevant environment.
- **TRL 7:** System prototype demonstration in operational environment (**CFD, Visualization, Process Modeling**) - Assist in demonstrating full scale pre-commercial system is demonstrated in operational environment, verify and validate the integration of upstream and downstream technologies
- **TRL 8:** System complete and qualified (**CFD, Visualization, Process Modeling, ML, ROM, Digital Twins**) - Assist in finalizing the technology for deployment and stable operation, training and maintenance documentation, and integration at system level
- **TRL 9:** System proven in an operational environment (**CFD, Visualization, Process Modeling, ML, ROM, Digital Twins**) - Assist in proving the technology fully operational and ready to be commercialized, system optimization for full-rate production.



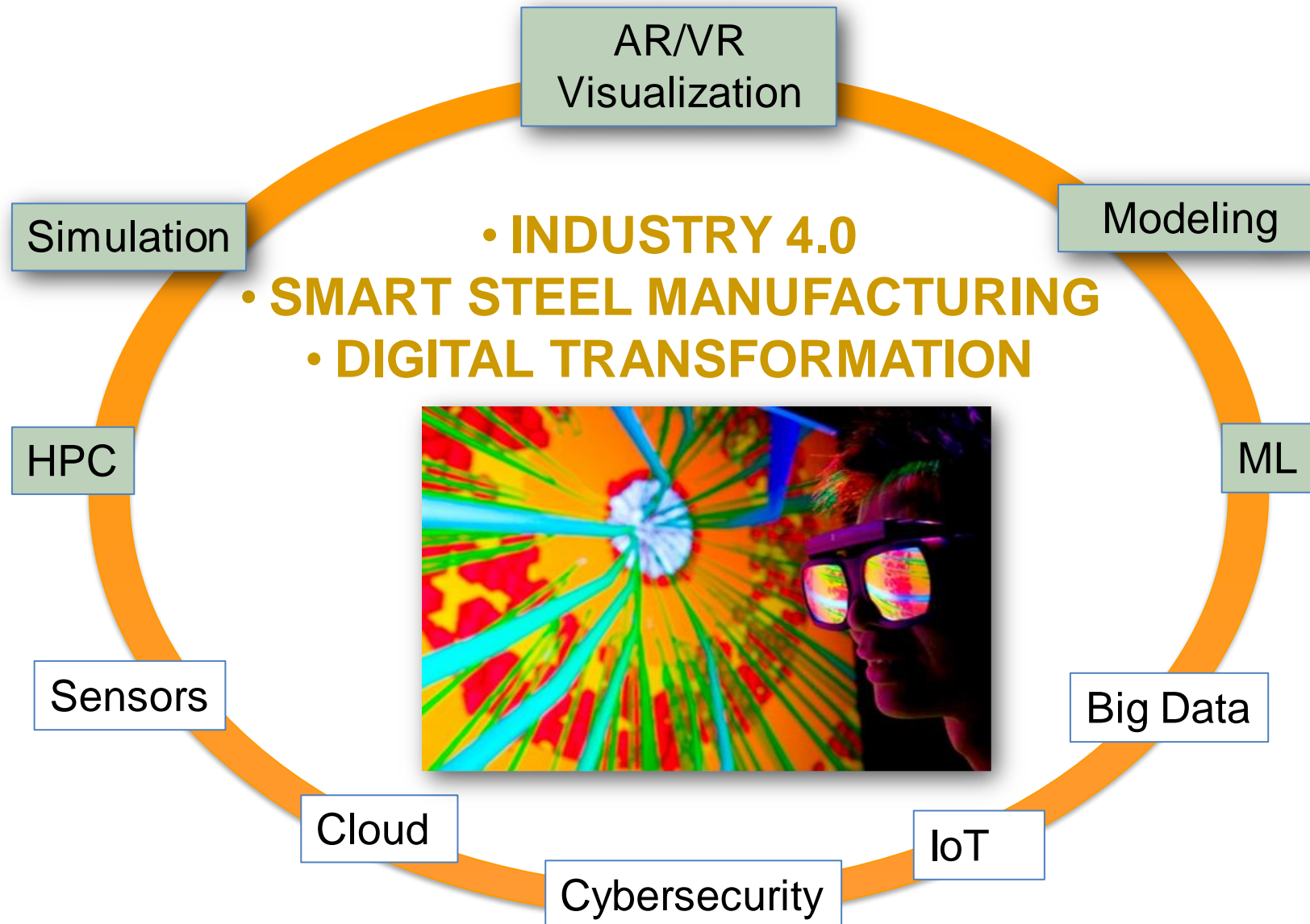
# CHALLENGES AND OPPORTUNITIES

- Culture evolution
- Connection to practical, real-world applications
- Physics-based, data driven, rapid predictions
- Big data (quality and completeness)
- Model validations
- Integration
- Collaboration
- Workforce development



**Advanced digital technology plays a key role in developing transformational technologies for Zero-emission Iron and Steelmaking!**

# FUTURE STEELMAKING





# THANK YOU!

- All the government funding agencies
- All the company sponsors
- All the collaborators
- SMSVC member companies
- CIVS faculty, staff, and students

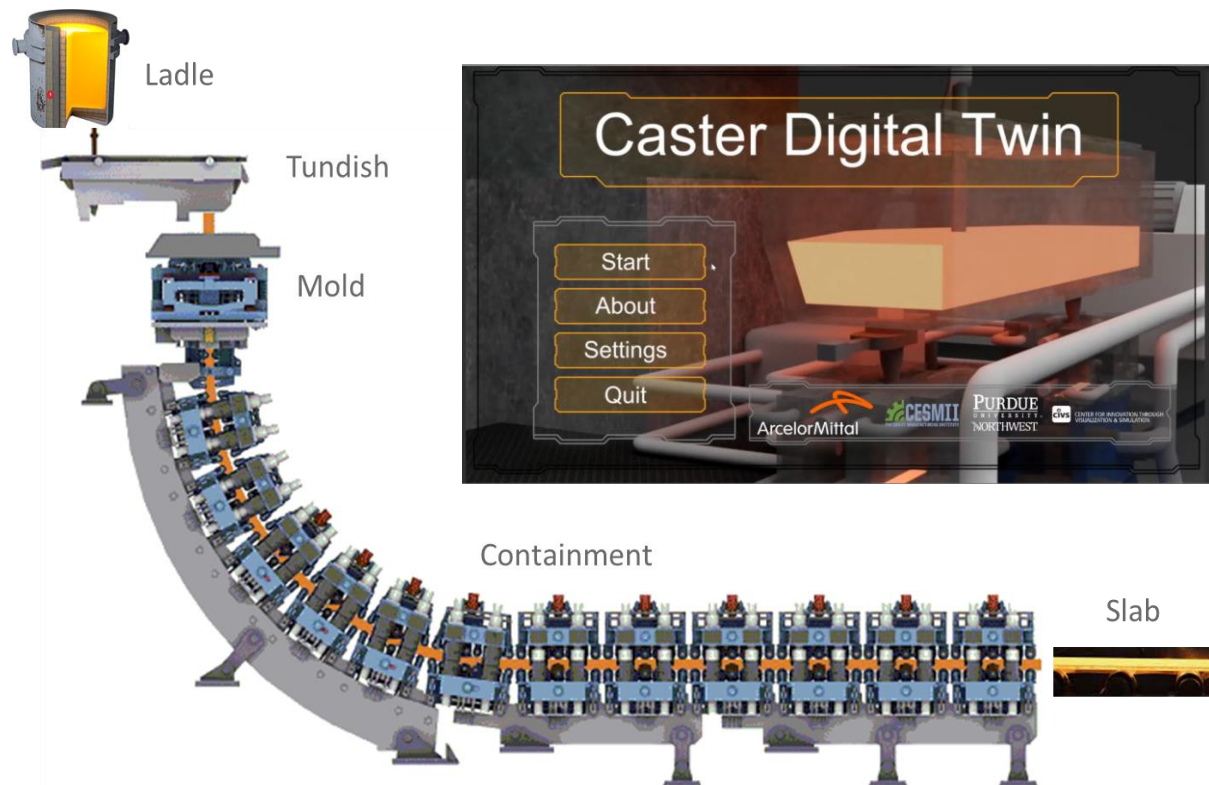
[civs@pnw.edu](mailto:civs@pnw.edu)

[centers.pnw.edu/civs](http://centers.pnw.edu/civs)

[www.steelconsortium.org](http://www.steelconsortium.org)

# 3-D CASTER DIGITAL TWIN

- **Objective:** To develop and deploy 3-D digital twin of a continuous casting process for producing steel slabs both offline and online
  - Track casting parameters relevant to slab quality
  - Track equipment, sensor and modeling parameters relevant to machine condition
- **Sponsor & Partners:** CESMII, Arcelormittal, Cleveland-Cliffs, PNW



## Digital Twin

- Digital replica of an actual physical caster
- Track important aspects of a process real-time and retrospectively
- Integration of real-time sensor-generated data, IoT, physics-based models (e.g. CFD), AI, machine learning and software analytics with 3D graphics tools
- Continuously update and change during operation
- Feedback to the control system

# PROTOTYPE CASTER DIGITAL TWIN



## OPERATING CONDITIONS

- Heat number
- Steel grade
- Casting speed, spray distribution
- Ladle weight
- Superheat

## SENSOR DATA

- Bearing cooling
- Mold cooling
- Machine Cooling

## CASTER OPERATION

- Operators and engineers
- Caster control room
- Maintenance



## MODEL DATA

- Sarclad data
- Slab quality tracking (GCM, CSQ, Dimensions)
- CFD

## VISUALIZATION

- Caster Health Monitor
- Off-line data stepping
- Sarclad data graphical display

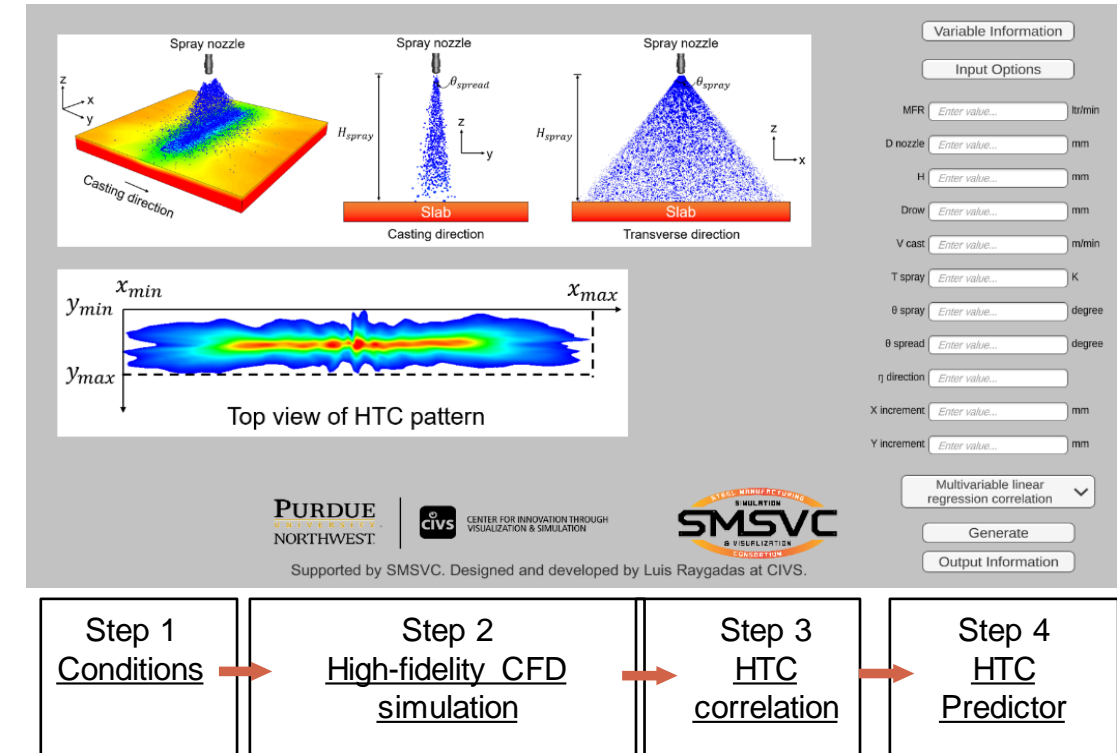


## ➤ Motivations:

- Dynamic control of spray water rate for continuous casting is critical to enhance steel strength.
- Heat transfer coefficient (HTC) distribution under each nozzle is required as boundary condition.
- HTC is often correlated with nozzle parameters & operating conditions.
- Experiment-based correlation development is labor-intensive & expensive.

## ➤ HTC Predictor:

- Multi-selection of HTC correlations
- Customizable correlation coefficients based on a specific caster

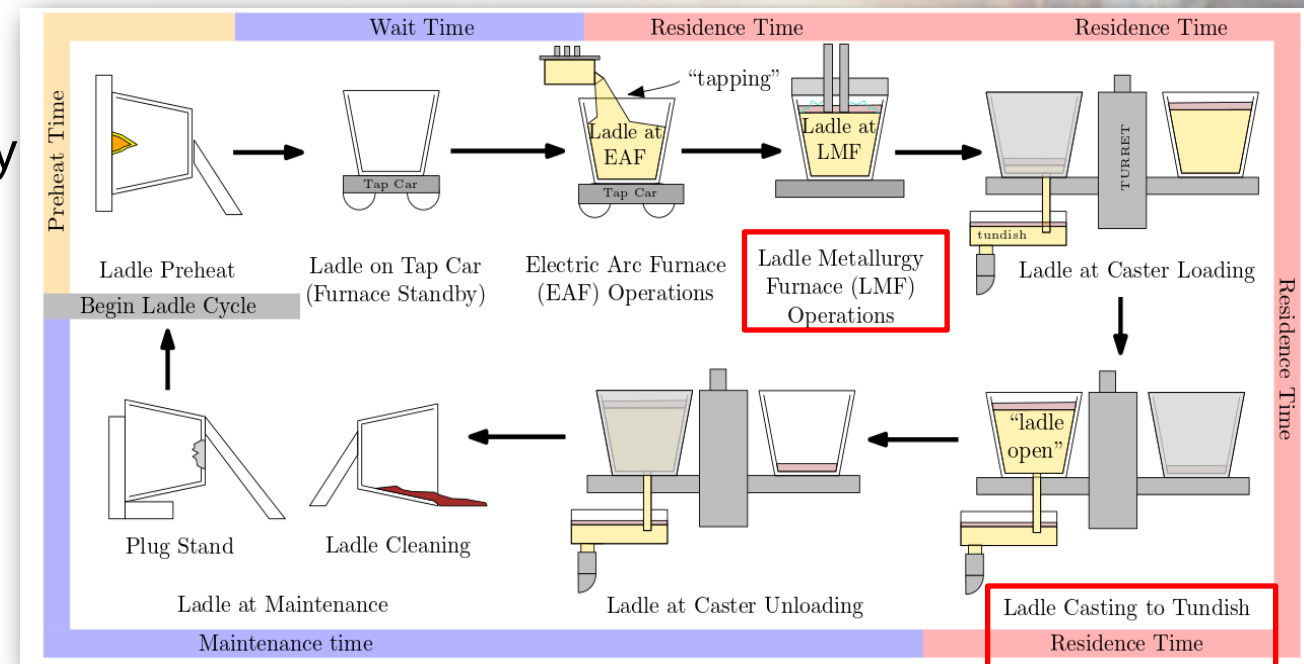
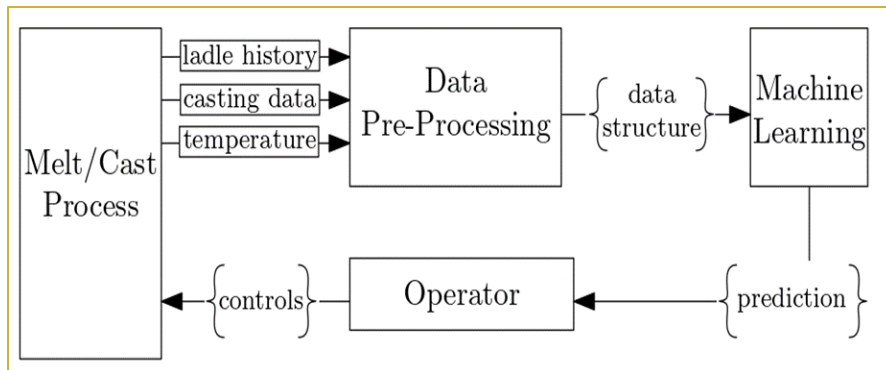


## Applications

- On-line real-time dynamic spray control
- Off-line/on-line malfunctioning nozzle identification
- Off-line nozzle selection
- Off-line researcher/operator training

# SMART LADLE

- **Process:** Ladle is used for transporting liquid steel from EAF or BOF to caster and for steel refining
- **Current Issue:** Casting temperature control by adjusting LMF temperature by operator experience
- **Objectives:**
  - To develop an AI tool for ladle operation
    - To accurately correlate casting temperature with known input variable conditions
    - To improve operational responses for casting temperature control
  - To develop generalized procedures for applying deep learning to other steelmaking processes
- **Impacts:**
  - Consistent casting quality
  - Increased productivity and energy efficiency
- **Sponsors:** AIST & SMSVC



## Example

- Steel Dynamics, Inc. Butler Division, IN
- Integrated with on-site production database for operator control interface
- Accuracy of predictions meets industry needs
  - Current average error of **3°F**, which is better than desired (5°F)

